

Q1. $P(H | \text{biased}) = 0.6$ $P(H | \text{fair}) = 0.5$
 $P(\text{biased}) = 0.5$

$$P(\text{biased} | H_{\text{twice}}) = \frac{P(H_{\text{twice}} | \text{biased}) P(\text{biased})}{P(H_{\text{twice}})}$$

$$= \frac{(0.6 \times 0.6) \times 0.5}{P(H_{\text{twice}})}$$

$$P(H_{\text{twice}}) = P(H_{\text{twice}} | \text{biased}) P(\text{biased}) + P(H_{\text{twice}} | \text{fair}) P(\text{fair})$$

$$= 0.5 (0.6^2 + 0.5^2)$$

$$\Rightarrow P(\text{biased} | H_{\text{twice}}) = 0.59 \quad \text{Ans}$$

Q2. $P(B) + P(G) = 1$ $P(G) = 0.9 \Rightarrow P(B) = 0.1$

$$P(SB | B) = 0.75 \quad P(SG | G) = 0.25$$

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$$P(B | SB) = \frac{P(SB | B) P(B)}{P(SB)}$$

$$= \frac{0.75 \times 0.1}{0.75 \times 0.1 + 0.25 \times 0.9}$$

$$= 0.25 \quad \text{Ans}$$

Events
 $B \rightarrow$ Car is blue
 $G \rightarrow$ Car is green
 $SB \rightarrow$ Car is seen as blue
 $SG \rightarrow$ Car is seen as green

$$P\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) = \frac{1}{(2\pi)^{n/2} |\Sigma|^{-1/2}} \exp\left(-\frac{1}{2} \left(\begin{bmatrix} x \\ y \end{bmatrix} - \begin{bmatrix} \mu_x \\ \mu_y \end{bmatrix}\right)^T \begin{bmatrix} \gamma_{xx} & \gamma_{xy} \\ \gamma_{yx} & \gamma_{yy} \end{bmatrix} \left(\begin{bmatrix} x \\ y \end{bmatrix} - \begin{bmatrix} \mu_x \\ \mu_y \end{bmatrix}\right)\right)$$

where $\begin{bmatrix} \gamma_{xx} & \gamma_{xy} \\ \gamma_{yx} & \gamma_{yy} \end{bmatrix}^{-1} = \begin{bmatrix} \Sigma_{xx} & \Sigma_{xy} \\ \Sigma_{yx} & \Sigma_{yy} \end{bmatrix}$

$$P(x|y=y_0) \propto p(x, y=y_0)$$

$$\propto \exp\left(-\frac{1}{2} (x-\mu_x)^T \gamma_{xx} (x-\mu_x) - (x-\mu_x)^T \gamma_{xy} (y_0-\mu_y) - \frac{1}{2} (y_0-\mu_y)^T \gamma_{yy} (y_0-\mu_y)\right)$$

$$\propto \exp\left(-\frac{1}{2} (x-\mu_x)^T \gamma_{xx} (x-\mu_x) - (x-\mu_x)^T \gamma_{xy} (y_0-\mu_y)\right)$$

$$\propto \exp\left(-\frac{1}{2} (x-\mu_x)^T \gamma_{xx} (x-\mu_x) - (x-\mu_x)^T \gamma_{xx} \gamma_{xx}^{-1} \gamma_{xy} (y_0-\mu_y)\right)$$

$$- \frac{1}{2} \left[(y_0-\mu_y)^T \gamma_{yx} \gamma_{xx}^{-1} \gamma_{xx} \gamma_{xx}^{-1} \gamma_{xy} (y_0-\mu_y) \right. \\ \left. + \frac{1}{2} \left[(y_0-\mu_y)^T \gamma_{yx} \gamma_{xx}^{-1} \gamma_{xx} \gamma_{xx}^{-1} \gamma_{xy} (y_0-\mu_y) \right] \right)$$

$$\propto \exp\left(-\frac{1}{2} \left(x-\mu_x + \gamma_{xx}^{-1} \gamma_{yx} (y_0-\mu_y)\right)^T \gamma_{xx} \left(x-\mu_x + \gamma_{xx}^{-1} \gamma_{yx} (y_0-\mu_y)\right)\right)$$

$$\cdot \exp\left(\frac{1}{2} (y_0-\mu_y)^T \gamma_{yx} \gamma_{xx}^{-1} \gamma_{xx} \gamma_{xx}^{-1} \gamma_{xy} (y_0-\mu_y)\right)$$

$$\propto \exp\left(-\frac{1}{2} \left(x-\mu_x + \gamma_{xx}^{-1} \gamma_{yx} (y_0-\mu_y)\right)^T \gamma_{xx} \left(x-\mu_x + \gamma_{xx}^{-1} \gamma_{yx} (y_0-\mu_y)\right)\right)$$

∴ From form of exponential in gaussian

for $p(x|y=y_0)$ $\mu' = \mu_x - \gamma_{xx}^{-1} \gamma_{xy} (y_0 - \mu_y)$; $\Sigma' = \gamma_{xx}$

from properties of inversion of block matrices $\begin{bmatrix} A & B \\ C & D \end{bmatrix}$

$$\mu' = \mu_x - \Sigma_{xy} \Sigma_{yy}^{-1} (y_0 - \mu_y) \Rightarrow \mu_x = \mu_x - \frac{\sigma_{xy}}{\sigma_y^2} (y - \mu_y)$$

$$\Sigma' = \Sigma_{xx} - \Sigma_{xy} \Sigma_{yy}^{-1} \Sigma_{yx}$$

$$\Rightarrow \sigma_x^2 = \sigma_{xx} - \frac{\sigma_{xy}^2}{\sigma_{yy}}$$

→ when x, y scalar variables

Q4 From Table 2,
 since on day 4, sensor ~~says~~ rainy,
 \therefore day 4 is indeed rainy

$$a = P(X_4 = R | Z_4 = R)$$

$$= P(Z_4 = R | X_4 = R) \overbrace{P(X_4 = R)}^k \eta$$

$$= 1 \cdot k$$

\downarrow
 prior belief

$X_d \rightarrow$ state of day d
 $Z_d \rightarrow$ measurement "
 R - Rainy
 C - Cloudy
 S - Sunny

$$b = P(X_4 = S | Z_4 = R)$$

$$= P(Z_4 = R | X_4 = S) \times k$$

$$= 0 \cdot k = 0$$

Similarly $c = P(X_4 = C | Z_4 = R) = 0$

$$\therefore a + b + c = 1 \Rightarrow k = 1$$

$$\Rightarrow P(X_4 = R | Z_4 = R) = 1$$

\downarrow
 posterior belief for day 4

for day 5

$$\bar{P}(X_5) = \sum P(X_5 | X_4) P(X_4)$$

$$\Rightarrow \bar{P}(X_5 = S) = 0.8 \times 0 + 0.4 \times 0 + 0.2 \times 1 = 0.2$$

$$\bar{P}(X_5 = C) = 0.6$$

$$\bar{P}(X_5 = R) = 0.2$$

posterior $\leftarrow P(X_5) = \eta P(Z_5 | X_5) \bar{P}(X_5)$

$$\Rightarrow P(X_5 = S) = \eta \times 0.6 \times 0.2 = 0.12\eta$$

$$P(X_5 = C) = 0.18\eta$$

$$P(X_5 = R) = 0$$

$$\sum P(X_5) = 1 \Rightarrow \eta = \frac{10}{3}$$

$$\therefore P(X_5 = S) = 0.4 \quad \text{Ans}$$

\downarrow
 prob it is indeed sunny

$$86 a) \quad x_t = ut - \frac{1}{2}gt^2$$

$$x_{t-1} = u(t-dt) - \frac{1}{2}g(t-dt)^2$$

$$\Rightarrow x_t = x_{t-1} + \dot{x}_{t-1}dt - \frac{g}{2}dt^2$$

$$\Rightarrow \begin{bmatrix} x \\ \dot{x} \end{bmatrix} = \begin{bmatrix} 1 & dt \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x_{t-1} \\ \dot{x}_{t-1} \end{bmatrix} + \begin{bmatrix} -\frac{g}{2}dt^2 \\ -gdt \end{bmatrix}$$

$$dt = 0.1 \text{ s}$$

$$\dot{x}_t = u - gt$$

$$\dot{x}_{t-1} = u - g(t-dt)$$

$$\Rightarrow \dot{x}_t = \dot{x}_{t-1} + (-g)dt$$